COST CURVES

- Directly plot the misclassif costs / error (in terms of prior probs)
- Might be easier to interpret than ROC, especially in case of different misclassif costs or priors

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Example:

- f₁ and f₂ with intersecting ROC curves
- f_2 dominates first, then f_1

BUT: Unclear for which thresholds, costs or class distribs f_2 better than f_1

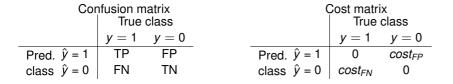
Nathalie Japkowicz (2004): Evaluating Learning Algorithms : A Classification Perspective. (p. 125)

ROC curves for f1 and f2

COST CURVES

Simplifying assumption: equal misclassif costs, i.e., $cost_{FN} = cost_{FP}$ \Rightarrow Expected misclassif cost reduces to misclassif error rate With law of total prob, we write error rate as function of π_+ :

$$\rho_{MCE}(\pi_{+}) = (1 - \pi_{+}) \cdot \mathbb{P}(\hat{y} = 1 | y = 0) + \pi_{+} \cdot \mathbb{P}(\hat{y} = 0 | y = 1)$$
$$= (1 - \pi_{+}) \cdot FPR + \pi_{+} \cdot FNR$$
$$= (FNR - FPR) \cdot \pi_{+} + FPR$$



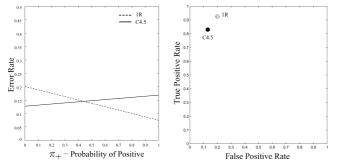


COST CURVES

• Cost line of a classifier with slope (*FNR* – *FPR*) and intercept *FPR*:

$$\rho_{MCE}(\pi_{+}) = (FNR - FPR) \cdot \pi_{+} + FPR$$

• Cost curves are point–line duals of ROC curves, i.e., a single classifier is represented by a point in the ROC space and by a line in cost space



Chris Drummond and Robert C. Holte (2006): Cost curves: An improved method for visualizing classifier performance. Machine Learning, 65, 95-130 (URL).

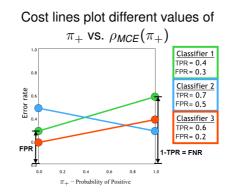


COST LINES

Cost line of a classifier with slope (FNR - FPR) and intercept FPR:

$$\rho_{MCE}(\pi_{+}) = (FNR - FPR) \cdot \pi_{+} + FPR$$

- Hard classifiers are points (TPR, FPR) in ROC space
- The cost line of a classifier connects (π₊, ρ_{MCE})-points at (0, FPR) and (1, 1 – TPR)
- Classifier 3 always dominates classifier 1
- Classifier 3 is better than classifier 2 when $\pi_+ < 0.7$

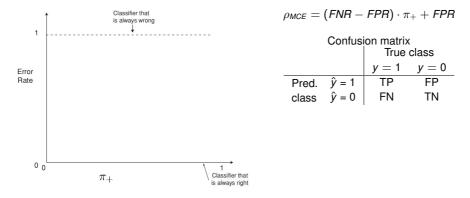


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COST LINES - EXAMPLE

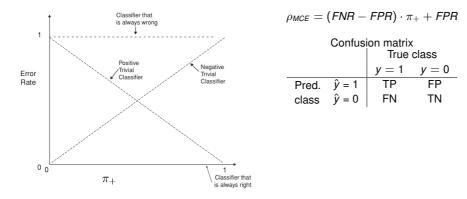
- Horizontal dashed line: worst classifier (100% error rate for all π_+) \Rightarrow *FNR* = *FPR* = 1
- x-axis: perfect classifier (0% error rate for all π_+) \Rightarrow *FNR* = *FPR* = 0

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COST LINES - EXAMPLE

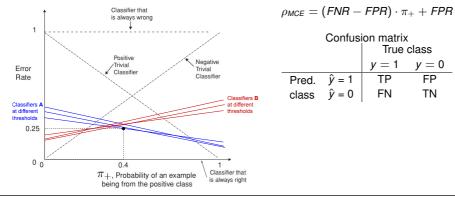
- Horizontal dashed line: worst classifier (100% error rate for all π_+) \Rightarrow *FNR* = *FPR* = 1
- x-axis: perfect classifier (0% error rate for all π_+) \Rightarrow *FNR* = *FPR* = 0
- Dashed diagonal lines: trivial classifiers, i.e., ascending diagonal always predicts negative instances (\$\Rightarrow FNR = 1\$ and FPR = 0) and vice versa



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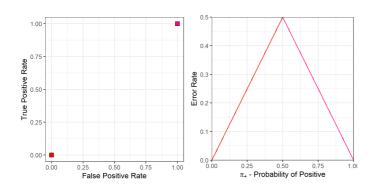
COST LINES - EXAMPLE

- Horizontal dashed line: worst classifier (100% error rate for all π_+) \Rightarrow *FNR* = *FPR* = 1
- x-axis: perfect classifier (0% error rate for all π_+) \Rightarrow *FNR* = *FPR* = 0
- Dashed diagonal lines: trivial classifiers, i.e., ascending diagonal always predicts negative instances (\$\Rightarrow FNR = 1\$ and FPR = 0) and vice versa
- Descending/ascending bold lines: two families of classifiers A and B (represented by points in their respective ROC curves)



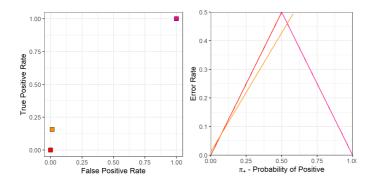
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- Left: ROC = TPR & FPR of a classifier for different prob thresholds
- Right: Corresponding cost lines
- Duality: For every ROC point we can construct the CC line, and vice versa.



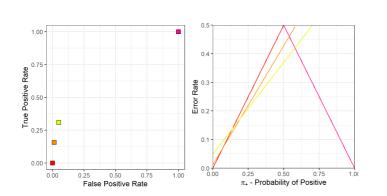
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- Left: ROC = TPR & FPR of a classifier for different prob thresholds
- Right: Corresponding cost lines
- Duality: For every ROC point we can construct the CC line, and vice versa.



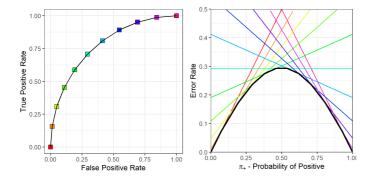
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- Left: ROC = TPR & FPR of a classifier for different prob thresholds
- Right: Corresponding cost lines
- Duality: For every ROC point we can construct the CC line, and vice versa.



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- Left: ROC = TPR & FPR of a classifier for different prob thresholds
- Right: Corresponding cost lines
- Duality: For every ROC point we can construct the CC line, and vice versa.
- Cost curve (right, black) is lower envelope of cost lines $\hat{=}$ pointwise minimum of error rate (as function of π_+)



0 0 X X 0 X X